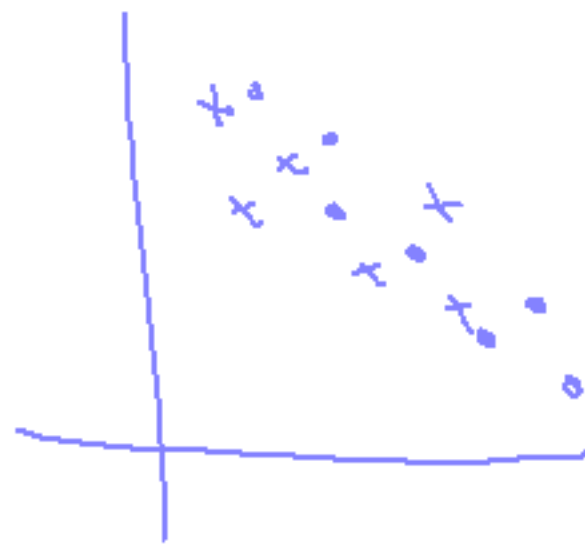


① Expl = X-var  $\leftarrow$  set/determined by researcher  
Resp = Y-var  $\leftarrow$  measured

Ex: mg. of a drug + # of tumors developed

② 2 quant. var's  
(#)

categorical var - color/shape

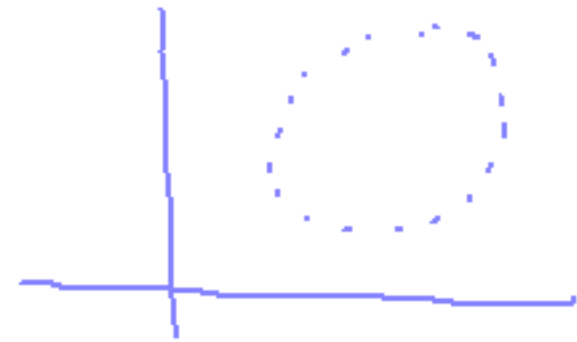


③  $r$  = correlation coefficient  
decimal measures strength of linear relationship

$$-1 \leq r \leq 1$$

$r = 0$  = no linear relat.

$r \approx -1, 1$  = strong linear



$r^2$  = coefficient of determination

% of the change in y-var.  
that was due to the change in  
the x-var.

④

⑤ Form: linear, curved, scattered

direction: + or - or scattered

Strength: strong  
moderate  
weak  
scattered

⑥ LSR =

$$\hat{y} = a + bx$$

LSR = least squares regression line

\* avg. of scatterplot

$$(\bar{x}, \bar{y})$$

8: LinReg(a+bx) X, Y, Y1

← puts line in  $y =$

← prediction

Y1(100)

$a =$

$b =$

$r =$

$r^2 =$

⑦ Residuals = errors btw. LSR line + actual data pt.

$$= \text{obs.} - \text{exp.}$$

↑  
actual data pt.

↑  
LSR line

Residual Plot:

x-var. vs. RESID

\* y-direction



line is not  
the best fit  
for data



RESID

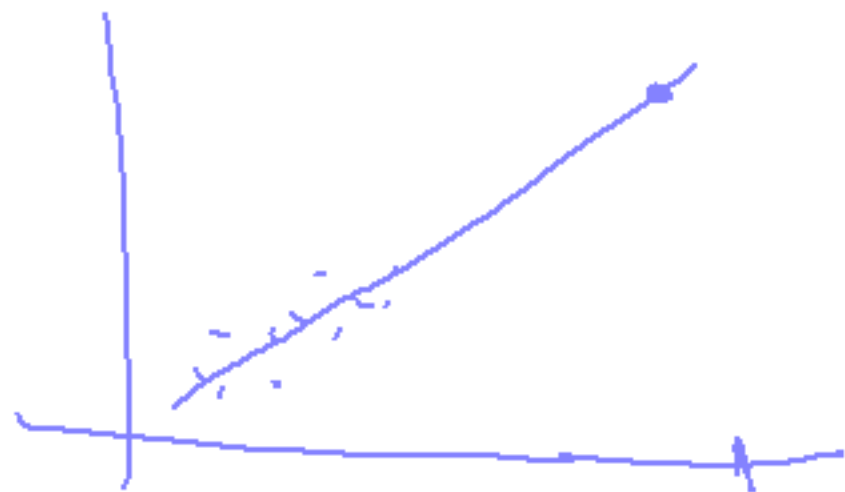


\* line is a good fit  
for data

Wkst - Airfares

⑧ For every 1 mile traveled,  
the airfare increased on  
avg. by \$0.12

⑪ extrapolation



sample 10.1 - testing the slope

LSR line

$$\hat{y} = a + bx$$

$\hat{p}$

$$a = \bar{y} - b\bar{x}$$

$$b = r \frac{s_y}{s_x}$$

Book/Form sheet

$$\hat{y} = b_0 + b_1 x$$

$$b_0 = \bar{y} - b_1 \bar{x}$$

$$b_1 = r \frac{s_y}{s_x}$$

stats  
sample

pop.  
param

$\hat{y}$

$y$

$a$

$\alpha$

pop. slope

$b$

$\beta$

$\beta_0$   
 $\beta_1$

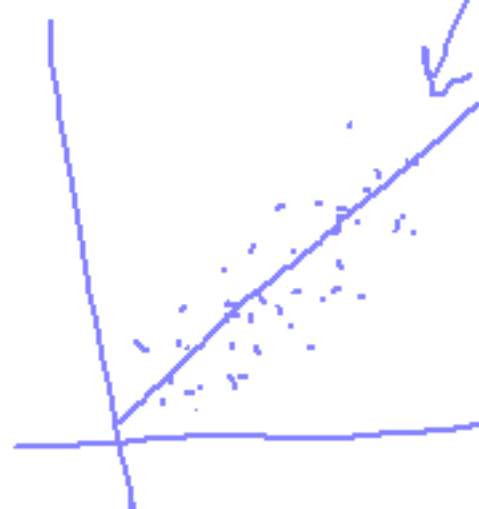
$e_i$

$\epsilon_i$

errors  
sample

LSR

pop. line  
of best  
fit





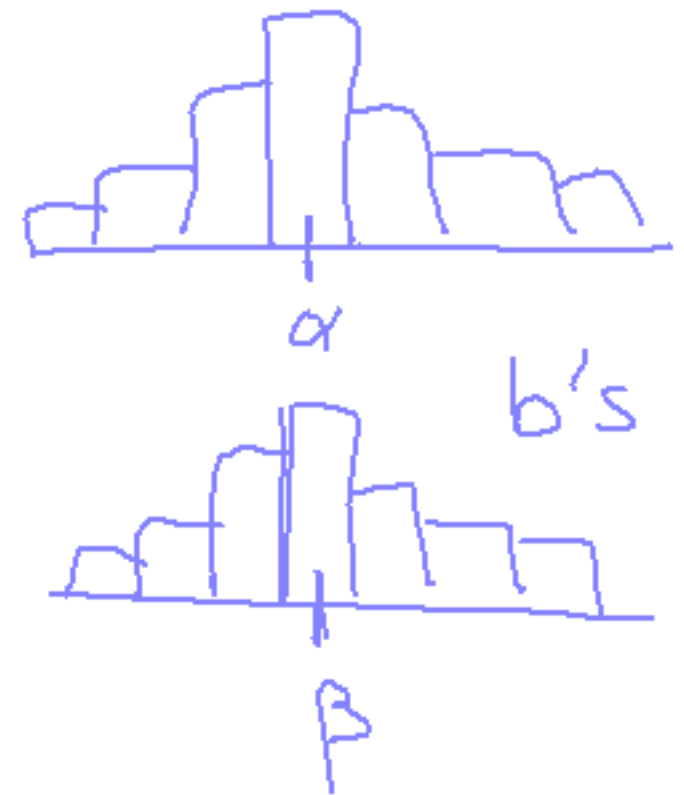
sample  
↓  
LSR line basis for  
inference about a population  
where our data is a sample.

LSR line:  $\hat{y} = a + bx + e_i$

Pop. line:  $y = \alpha + \beta x + \epsilon_i$

a and b

- unbiased estimators of  $\alpha$  and  $\beta$  a's
- normally distributed centers @  $\alpha$  and  $\beta$



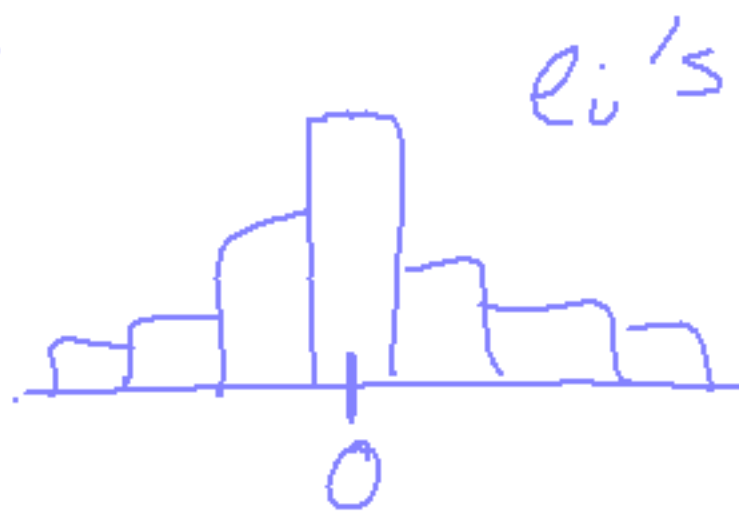
$E_i$  = residuals = deviations

- estimated by  $e_i$
- independent
- normally distrib.
- mean = 0

std. dev. of residuals =  $\sigma$

- resid = obs. - predicted

- $\sum e_i = 0$



pop.  
↓  
 $\sigma$  = std. dev. of resid.

• estimated by  $S$

resid  
•  $S = \sqrt{\frac{\sum e_i^2}{n-2}}$

•  $df = n - 2$

estimate of  $\sigma = a$   
 $\beta = b$

